



Climate extremes: Data issues

Lisa Alexander

Climate Change Research Centre and ARC Centre of Excellence for Climate Systems Science, University of New South Wales

WCRP-ICTP 21st July 2014











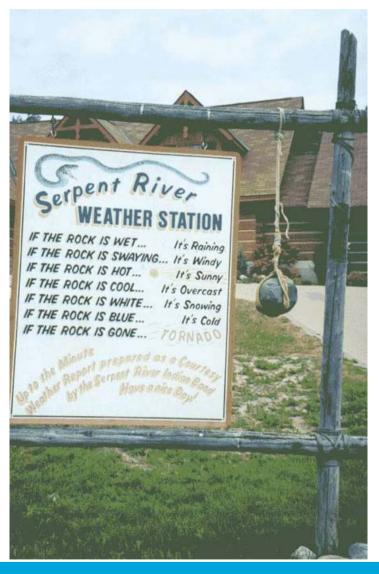
WCRP Grand Challenge on Extremes

 8 key scientific questions have been posed to the scientific community over the coming decade

Q1: How can we improve the collation, dissemination and quality of observations needed to assess extremes and what new observations do we need?



Why observe the climate?





How do we observe the climate?



This is not the lab!

Michael de Podesta's 'Instrument of real beauty' for determining the Boltzmann constant



Observations in some in really challenging environments ...



Image courtesy BAS via CIMSS AWS website (Fossil Bluff station)



Many different observations









	rove.	moer	1795.				
	0.	2.	1. 1.	2.	8 Men	8.2.8	Remarks & Observations
,	XW	NW.	No Close	1 2.	2:		Moderate
2			48 66wa				O
3	N	A .	VW Elm	Chan	Elean		O.
4			w bles				A new shared .
5			sw blen				De Olin Gan Donder France .
6			5 Clear			-	OF.
7	SW	SWS	W French	Clear	Clear		D3 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
80	e SW	SW.	SW Char	auan	llear .		De la Proposition Maria 300 30
9	5.	34 -	VE Rain	Rain	Clan.		Heavy Gale _ Colon _ Fred Breeze
10	Nn	No.	5W bles	blear	blea		moderate - my france
11	5	5	5 Cloud	· Clear	640		D. 32 - 19 11 11 11
12	N	NE.	48 Clian	Glasia,	Rain		De State Cart Con
13	N	dr.	NE Blea	blean	blea		2° ()
14	KE:	3	SE Blea	· Clear	Char.		De Ship Grace arrived from Manty.
156	3 NE	WS	W Close	ble	Char	1	D. Ship Grace arrived from North
16	.6	6.	E Cloud	Dull	Rain		D. Brig Orange Carberry arrived 8
17	N	N.	Mr Dule	andy	blear		
18	NE	N.	Nor Glas	Clear	Char		2. Harry sailed my so
			sw blea				D' & layle you with the stope sail the to
20	Sir	m	W blond	, alond	Monday		De Chighe I, a the Broadlope of the Marine
21	514	SW	W Dule	Rain	blus,		
220	2 N	N.	N Chas	· andy	Blody		Fond Bries
23	NW	NW.	NW 66.0	y andy	Clear		Fred Bours - 1
24			5 au				moderate 1 1 100
26	NW	· NW.	NW bles	Clear	blan		Fred Bryer - Shineversary of line
26	NH	NW.	SW Bles.	Chan.	Cleany		The Brun - Anniversary of Plane of the Severn arised - General Thanksyung
27	N	NE.	N blea	· Celenty	Dell	10 100	modrate /









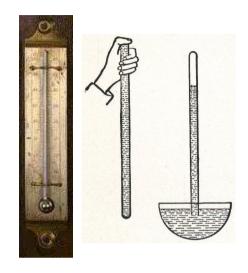






History of instrumentation and observations

- Between 16th and 17th C instruments started to be constructed to measure weather variables e.g. temperature (Galileo, Santorio), pressure (Torricelli)
- 18th and 19th century
 - > 1724 Farenheit introduces first standardised scale
 - ➤ 1742 Celcius introduces scale based on freezing and boiling point of water
 - ➤ 1848 Kelvin introduces scale based on 'absolute temperature'
- Military realises benefit of weather observations (first daily forecast produced -1860)







20th century

Late 1800s/Early 1900s – National Weather Services formed



1941 – first radar network

1959 – first weather satellite launched (Vanguard 2)

1975 – first GOES satellite launched









Data requirements for extremes

- Quality, consistency and availability of observations underpin analysis of climate extremes and attribution analyses
- Errors in data are likely to show up as 'extreme'
- But what we want/need versus what we get can be very different







Quality – what we want

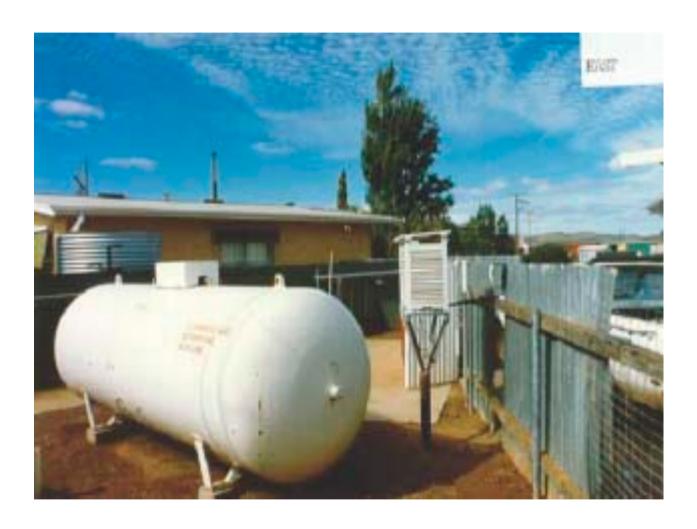








Quality – what we get



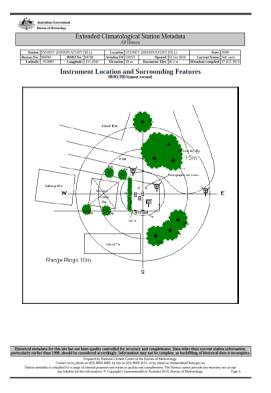


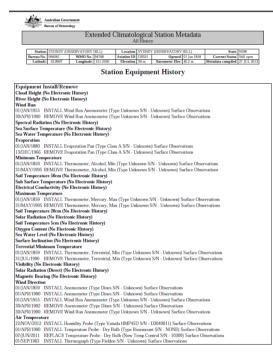


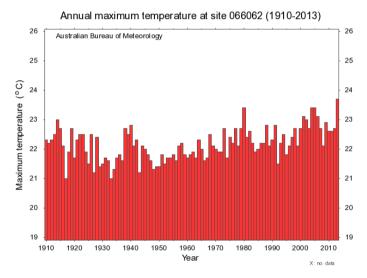


Consistency – what we want

Artificial changes are well documented (metadata) and have been accounted for over all time periods (traceability)







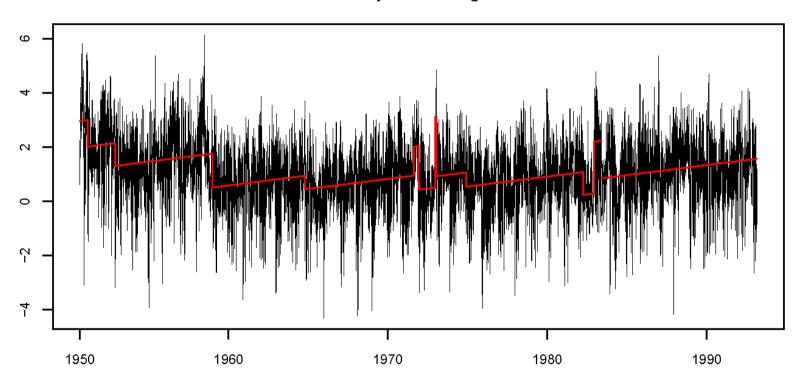






Consistency – what we get

Base anomaly series and regression fit

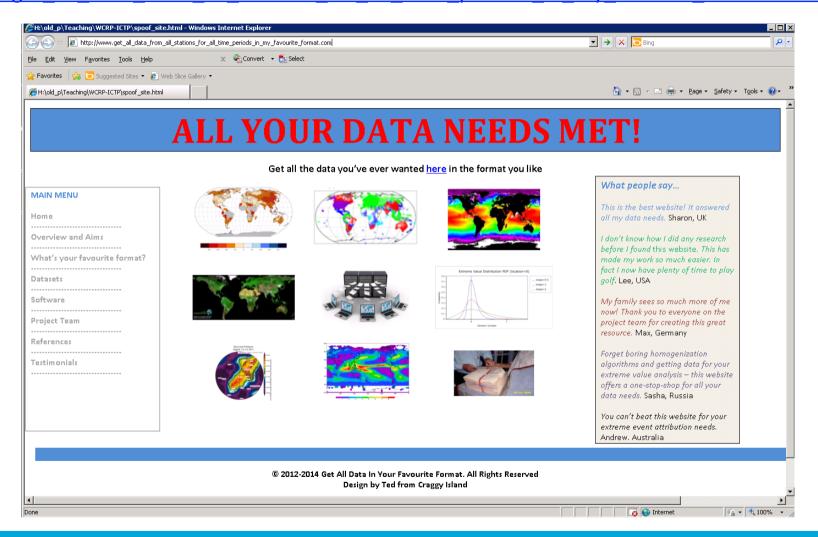






Availability – what we want

www.get all data from all stations for all time periods in my favorite format.com

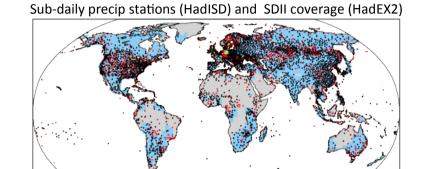








Availability – what we get







"We estimate the world loses 500,000 old records each day". Rick Crouthamel, IEDRO



years < 5



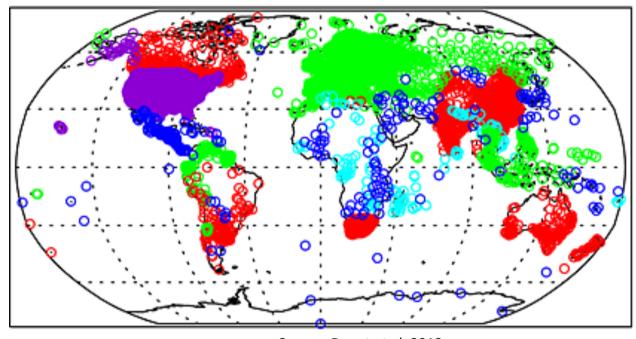


Issues and uncertainties – 1. station error



uncertainty of individual station measurements

Issues and uncertainties: 2. sampling error

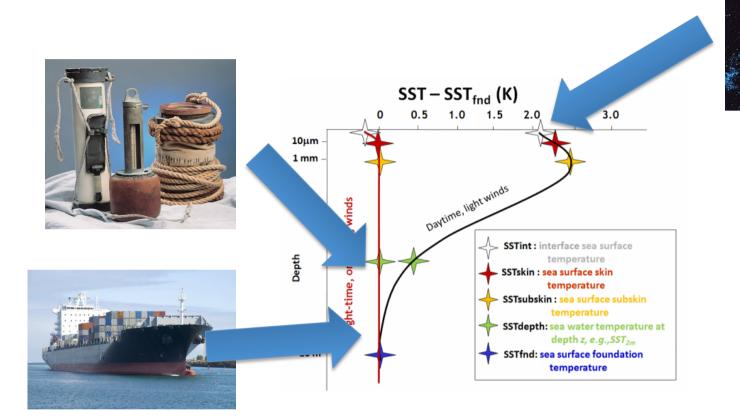


uncertainty in area mean caused by sampling small number of point values

Source: Donat et al. 2013



Issues and uncertainties: 3. bias error

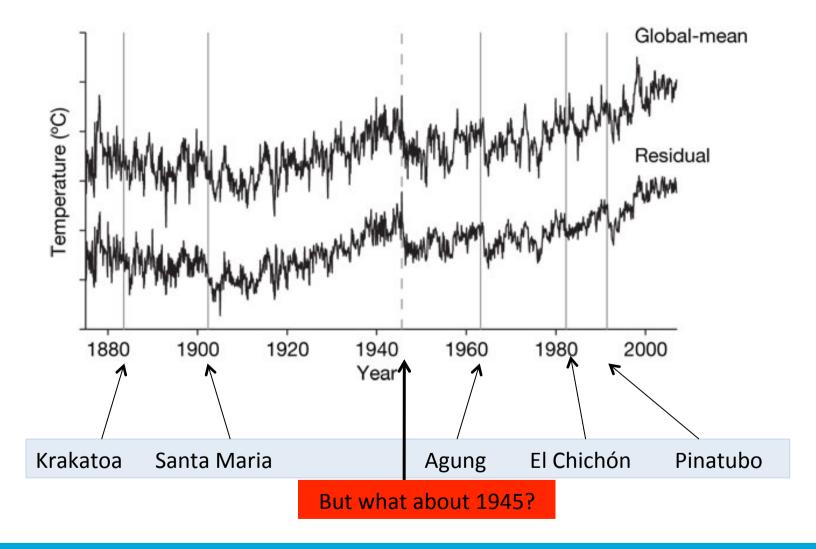


uncertainty in large-scale values caused by systematic changes in measurement methods



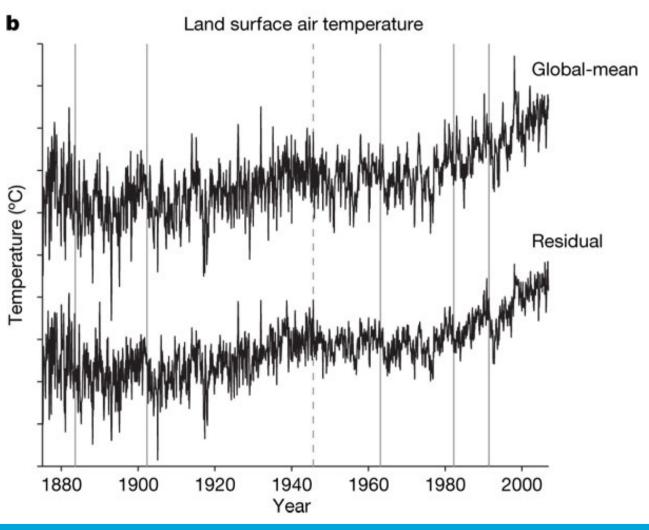


Global SST measurements

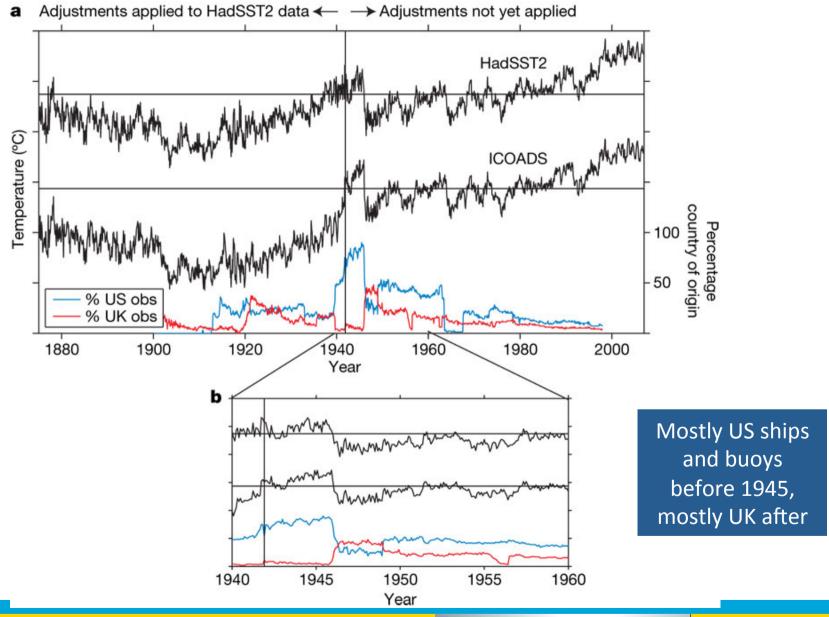




No evidence of change over land









Inhomogeneities

Artificial changes in the observed record that are not due to changes in climate

Can be introduced through changes in e.g.

observation source or observing practice

number of observations

station relocation

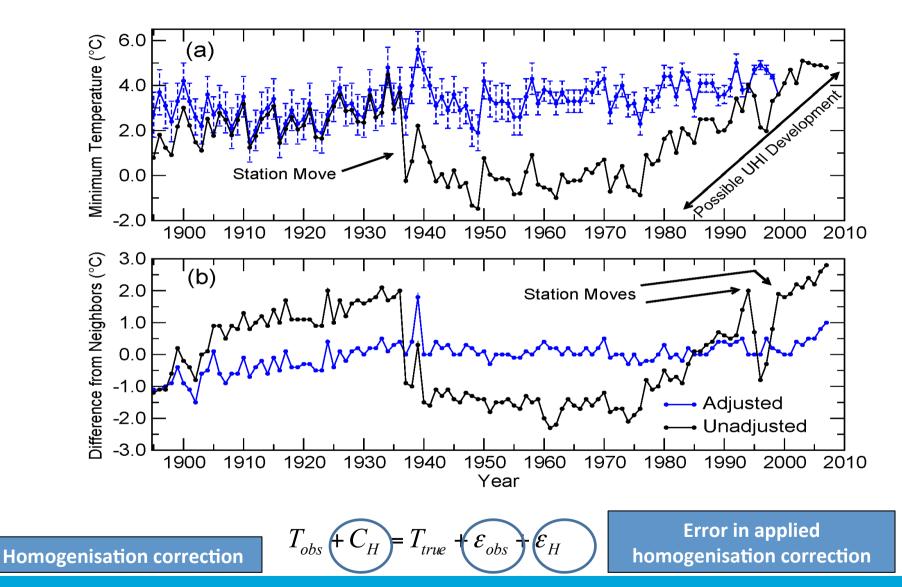
instrumentation change

site environment

Inhomogeneities can be sudden or gradual



Statistical tests can adjust data for inhomogeneities

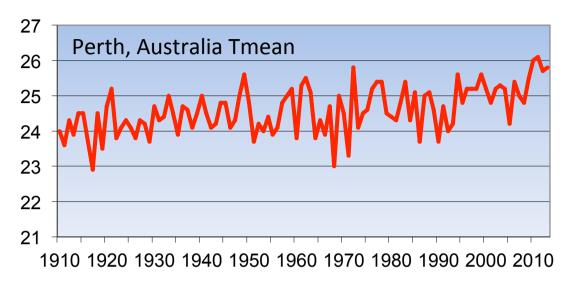


Random measurement error

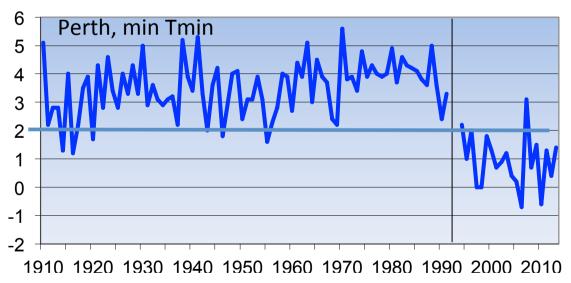




Is this different for extremes?

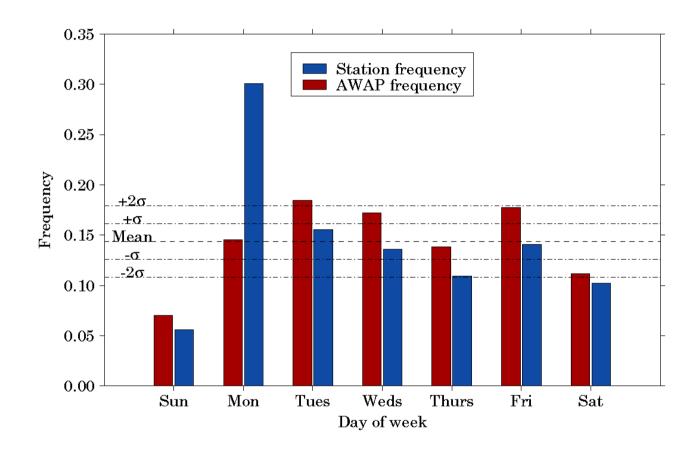


Before a site move in 1993 an 'extreme' Tmin < 2°C occurred only 5 times in Perth but almost every year since then





Observing practice – it never rains on a Sunday





The long list of issues to be considered

- Station moves
- Instrument changes
- Observer changes
- Automation
- Time of observation biases
- Microclimate exposure changes
- Urbanization
- And so on ...
- And that's just for the land records, similar laundry lists for satellites, radiosondes, marine ...



Underlying which are two absolutely fundamental issues ...

- A lack of traceability to absolute or even relative standards for most of the historical (and present day) records
- A lack of adequate documentation of the (ubiquitous) changes sufficient to characterize in an absolute sense the time-varying measurement characteristics.









All data have problems and uncertainties

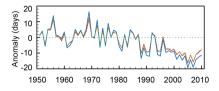
- Which are different from (climate) models
- Which require expertise to minimize
 - E.g. not all data sets were created equal
- Fortunately, there are experts in almost all types of observational data who would be happy to share their insights
 - As these data are often their life's work

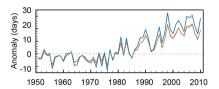




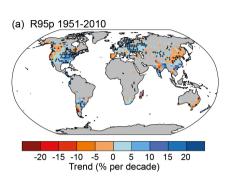


IPCC AR5 observed extremes assessment

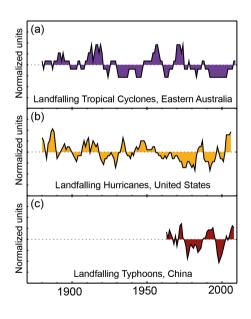




Very likely decreases in cold days and nights and increases in warm days and nights

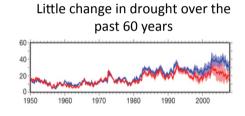


Likely more land regions where the number of heavy precipitation events has increased than where it has decreased

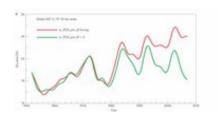


Low confidence in long-term increases in intense tropical cyclone activity

Virtually certain
North Atlantic since
1970s



Increasing drought under global warming in observations and models



Low confidence in increasing drought duration/intensity

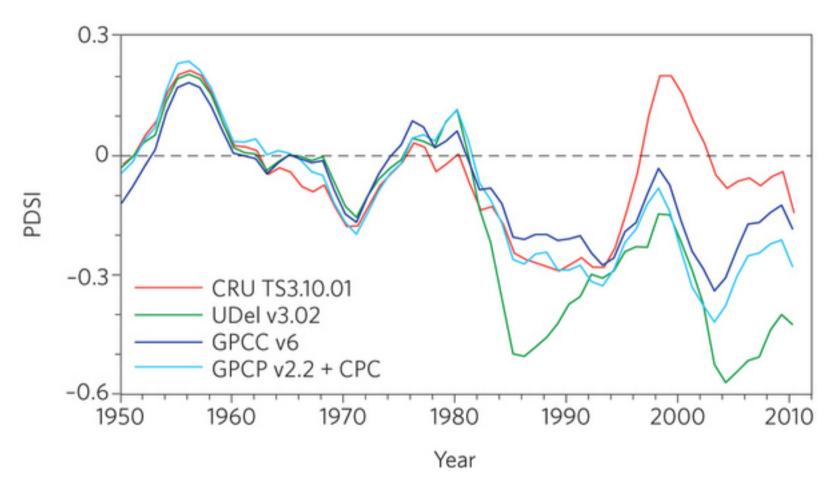
Likely increases and decreases in some regions







Recent attempts to reconcile differences



Source: Trenberth et al. 2014







Scaling issues

Fundamental mismatch?

 The spatial representativeness of in situ observations which are gridded using interpolation techniques may not 'mean' the same as climate model output of extremes

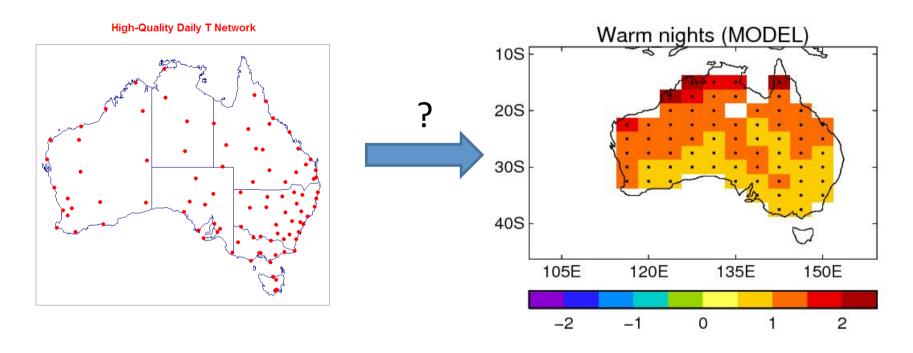
- Scale mismatch (or 'problem of change in support') more importantly affects phenomena whose spatial features are discontinuous or have short temporal scales
 - e.g. sub-daily precipitation, extreme events

 Alternative data sets are available (e.g. reanalyses) but come with their own problems



Points versus grids

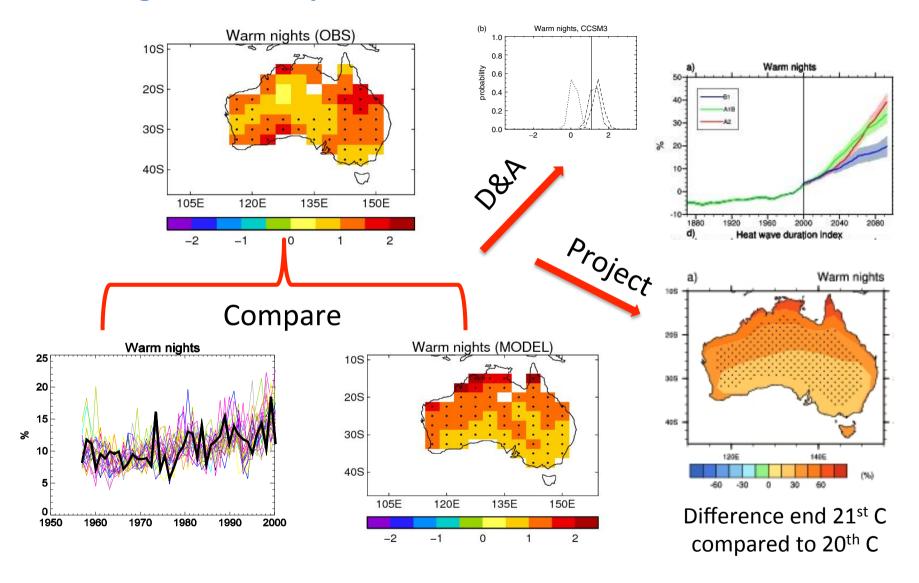
- 1. Observations are taken at points locations
- 2. Climate model output represents an areal average



How can we compare 1. and 2.?



Gridding allows comparison with climate models







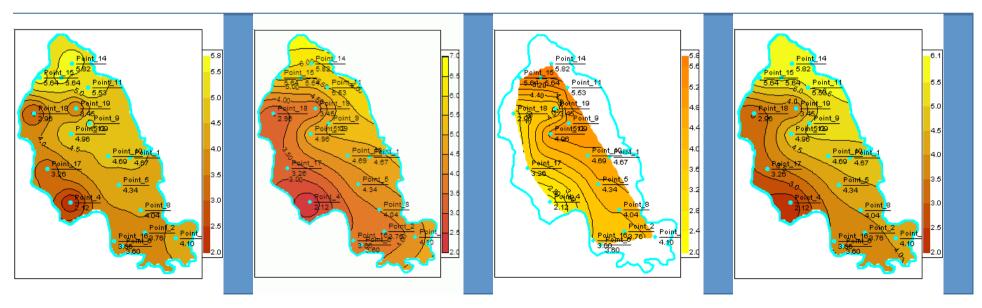
Gridding versus downscaling

- Gridding methods usually work by assigning values to unknown points by means of a weighted average of a number of known points
- Downscaling is the means of converting of areally averaged data to point data. There are two main types:
 - Statistical
 - Dynamical

Gridding methods

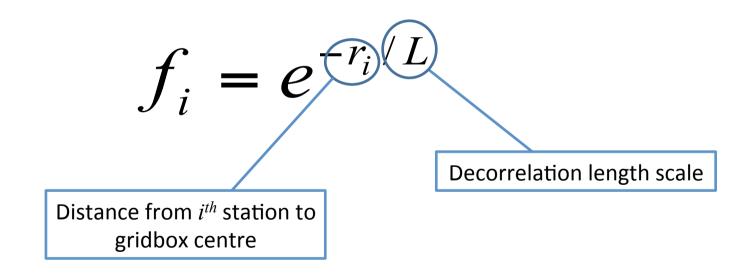
- There are many techniques that convert point observations to areal averages (grids). These include:
 - Kriging
 - Natural neighbours
 - Minimum curvature
 - Inverse distance weighting
 - Angular distance weighting
 - Radial Basis Functions

— ...



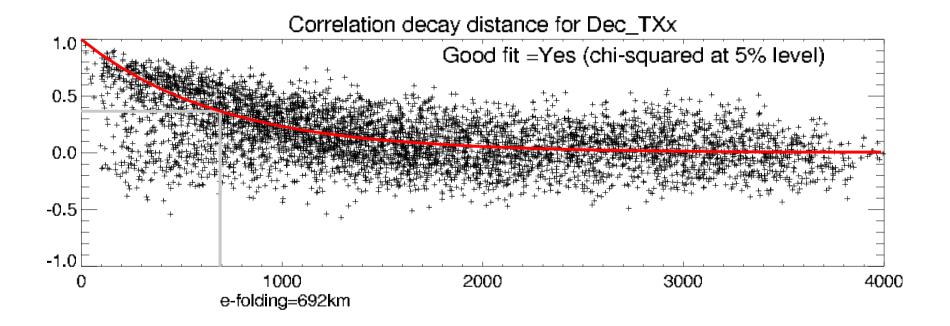
Example: Angular Distance Weighting (ADW)

- Requires knowledge of the correlation structure of the data i.e. how the chosen climate variable varies across space and time
- A correlation function, f, is defined for each i^{th} station:



Decorrelation length scale

- Also referred to as the e-folding distance
 - i.e. the time it takes the fitted function to fall away with distance to 1/e





Calculating weighting functions

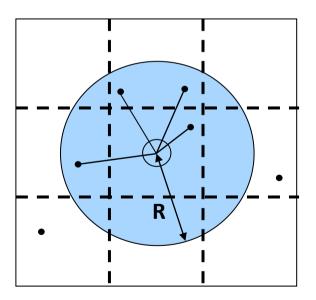
Bearing of i^{th} and k^{th} stations

$$\omega_{i} = f_{i}^{m} \left\{ 1 + \frac{\sum_{k} f_{k}^{m} \left[1 - \cos(\theta_{k} - \theta_{i}) \right]}{\sum_{k} f_{k}^{m}} \right\}, \quad i \neq k$$

Adjusts function decay

k sums over all stations within circle of influence (R)

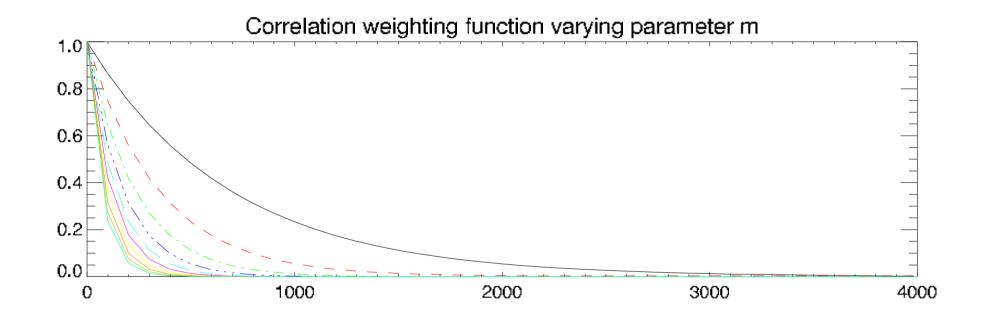
R is determined by the correlation function (usually the e-folding distance)





Testing values of m

 The parameter that determines the exponential decay of station weights

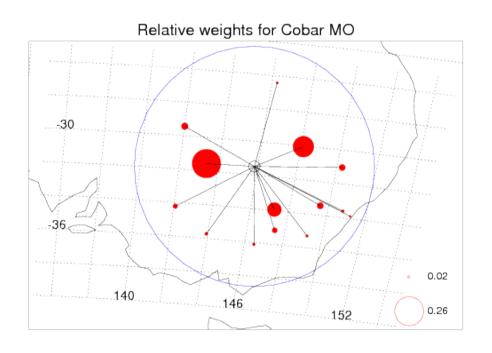


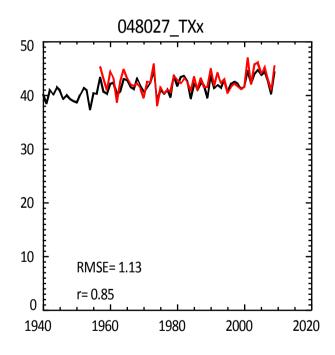
Parameter choices

- Period over which correlation is performed (all years, subset of years)
- Value of m (multiple tests can determine 'best' value)
- Distance over which function is fitted
- Ultimately parameter choices necessarily have to reflect compromises (don't have data points for all space and time periods, don't have all information)

Testing the 'goodness of fit' of the chosen method

 The method can be tested by trying to simulate the station data rather than an unknown grid point location

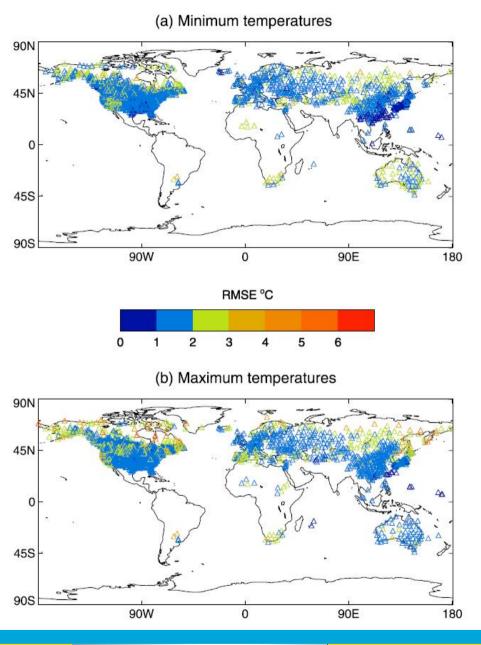






Error fields

- •Possible to determine regions where you have more/less confidence in the gridded product
- •Enables 'error bars' to be placed on gridded observational data
- •Important for climate model evaluation







Parametric versus structural uncertainties

- Parametric uncertainties can be calculated by testing our choices within our model framework e.g.
 - Parameter settings
 - Minimum number of 'points' to consider
 - Type of function to fit and maximum distance over which to fit function
 - Value of function decay
- Structural uncertainties can be calculated by using multiple gridding methods
 - Questioning your methodological framework
 - E.g. Kriging versus ADW
- Need to understand uncertainties before appropriate evaluation of climate models

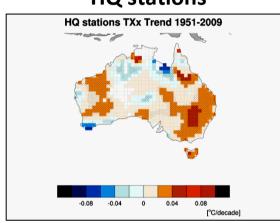


Comparison of the different data sets

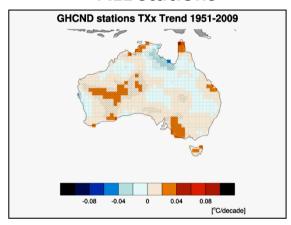
Hottest day of the year

(1° grid, fixed parameters)

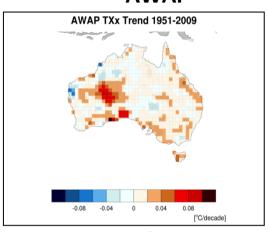
HQ stations



ALL stations



AWAP



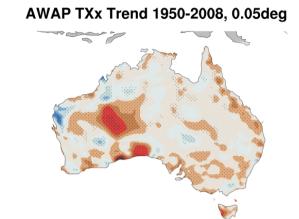
Same gridding method Different input data Same input data
Different gridding method



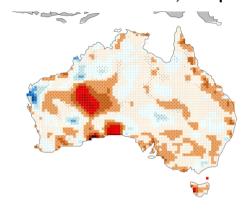




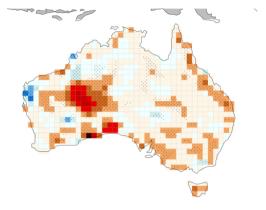
Also does grid size matter...?



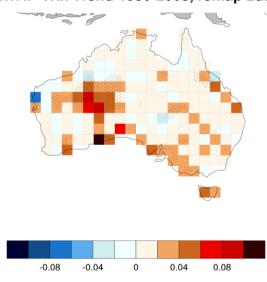
AWAP TXx Trend 1950-2008, remap 0.5deg



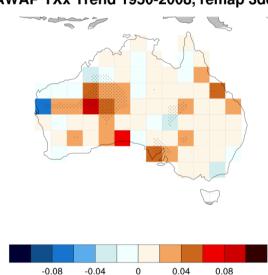
AWAP TXx Trend 1950-2008, remap 1deg



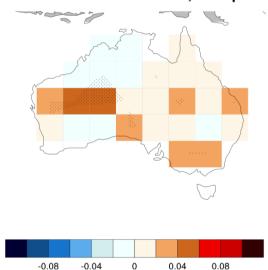
AWAP TXx Trend 1950-2008, remap 2deg



AWAP TXx Trend 1950-2008, remap 3deg



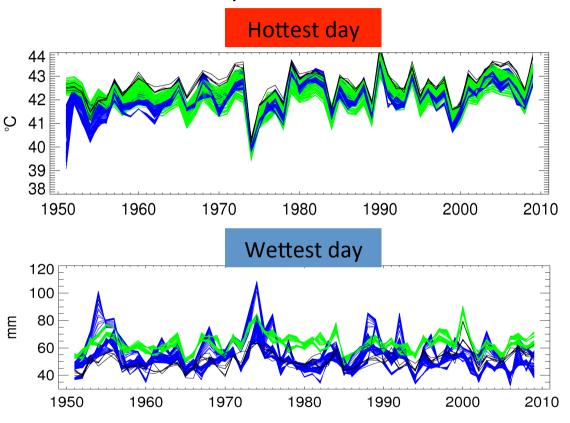
AWAP TXx Trend 1950-2008, remap 5deg





How large is this uncertainty?

Example for Australia



Several hundred realizations by varying:

- Grid size
- Parameters
- Network density
- Interpolation method
- Prior painstaking work on quality control and homogenisation of data



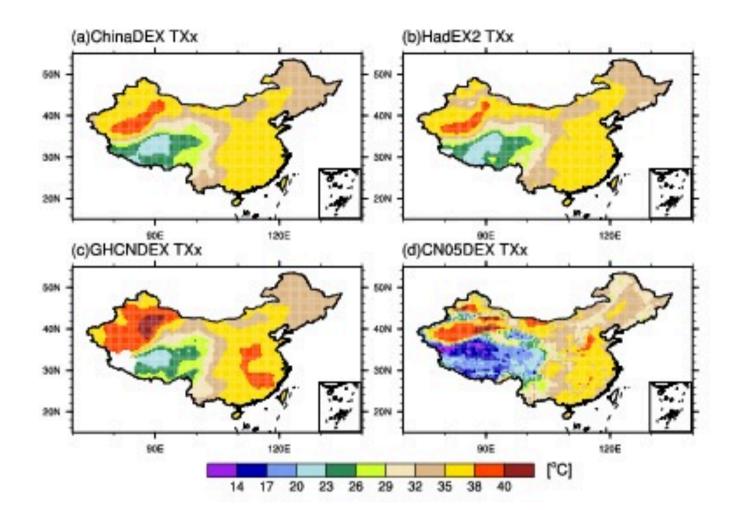
Source: Vogel, 2012







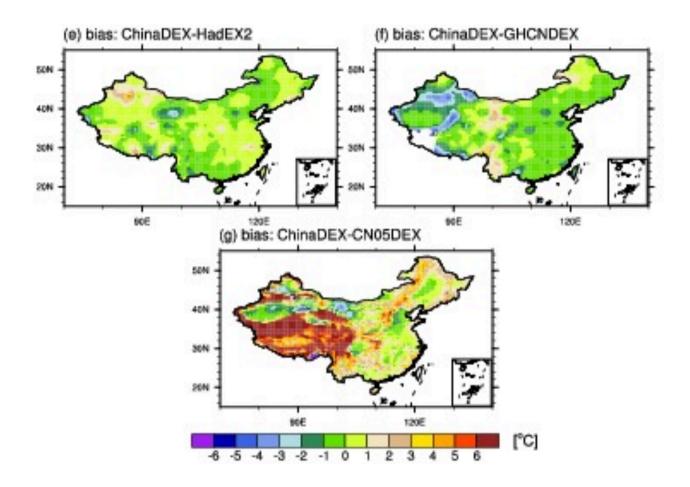
Can test multiple methods and regions





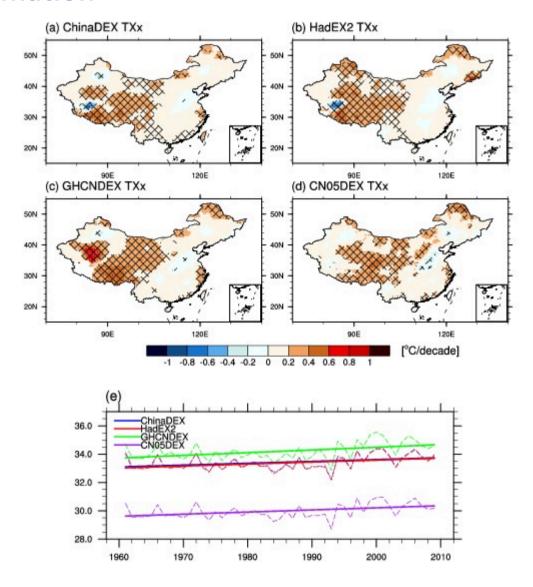


Bias estimation



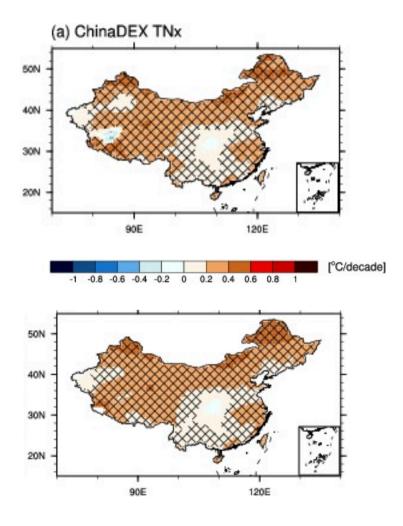


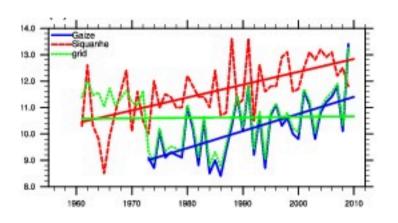
Trend estimation





For absolute extremes quality particularly important







Why don't we just use reanalysis data?

- Reanalyses based on observational data assimilated into a NWP model
 - ✓ provide physically consistent fields
 - ✓ complete spatial coverage
 - X Usually shorter than in situ records
 - X Inhomogeneous especially since assimilation of satellite records

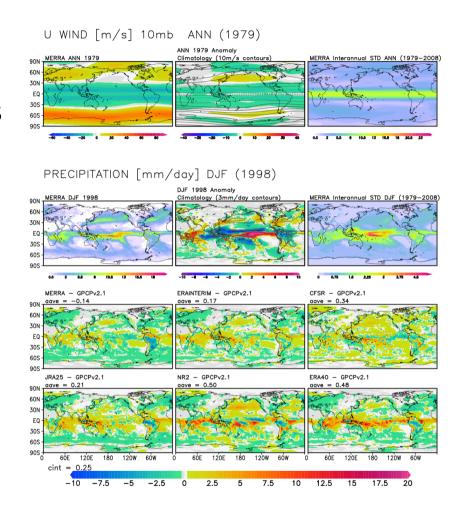






Advantages and disadvantages of reanalyses

- Able to assess fields for which there are little or no observations
- Output may be more similar to GCMs
- Disagreement between products and between products and observations
- Inhomogeneous over time (especially due to inclusion of satellite data in late 1970s)





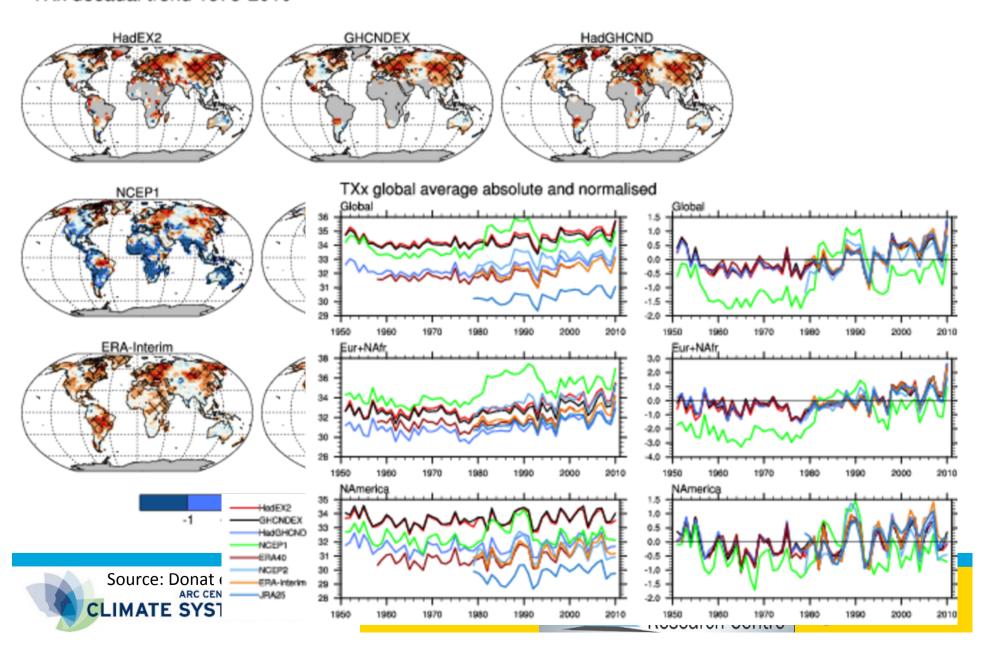
Types of reanalysis products

- European Centre for Medium-range Weather Forecasts (ECMWF) e.g.
 - ERA-40 (Sep 1957-Aug 2002)
 - ERA-interim (1979-present)
- USA e.g.
 - NCEP/NCAR reanalysis (1948/01/01 to present)
 - NASA MERRA (1979-present)
 - NOAA-CIRES 20th Century Reanalysis V2 (20CR): 1871-2008
- Japan e.g.
 - JRA-25 (1979-2004)

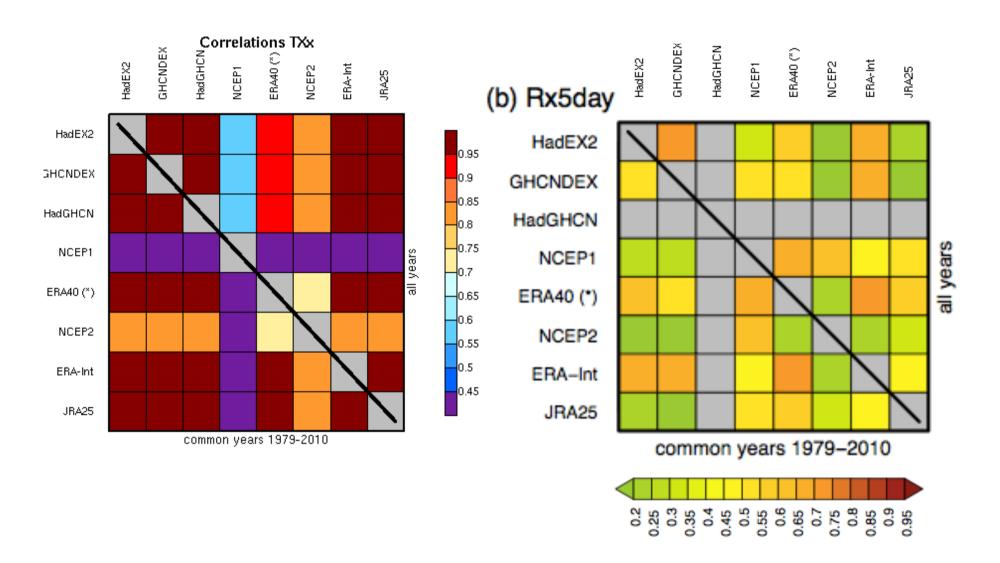


Reanalysis data in comparison to gridded in situ data sets

TXx decadal trend 1979-2010



Example spatial correlation between datasets



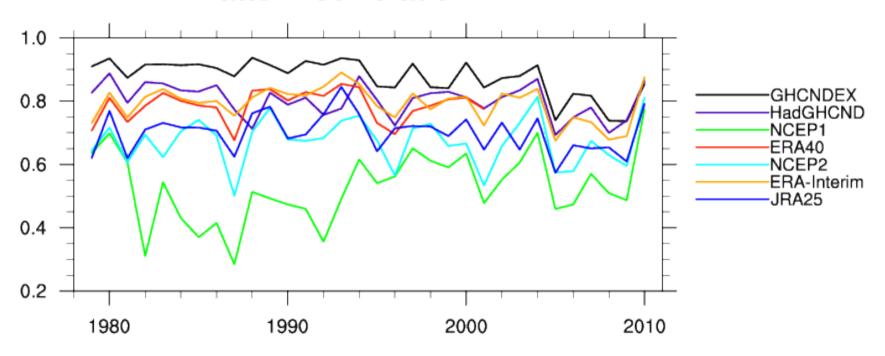






Example temporal correlation between datasets

Pattern correlation TXx

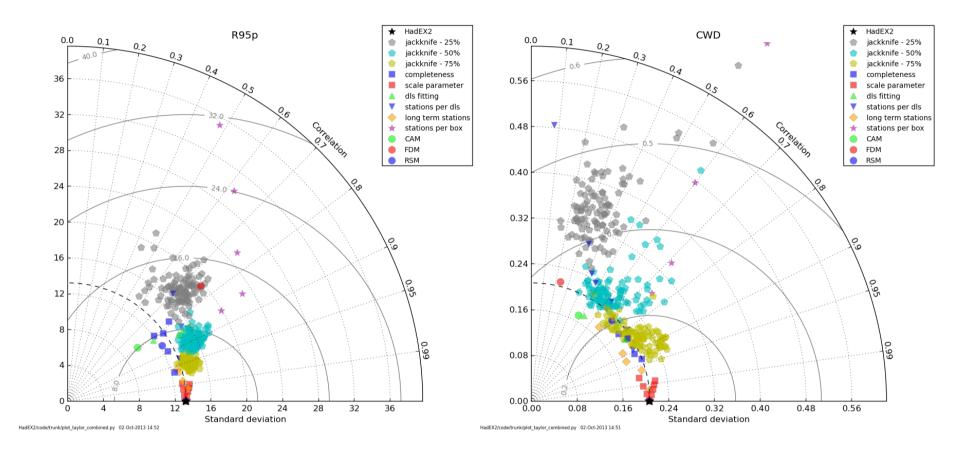








Calculating uncertainties for global datasets of extremes



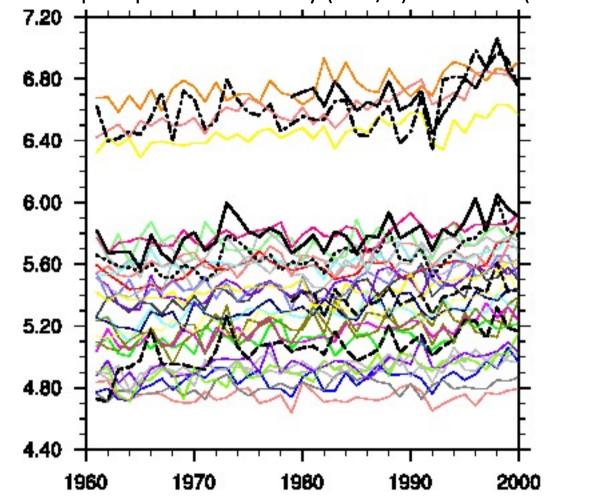
Structural uncertainties are larger than parametric uncertainties Some indices have larger uncertainties than others





'Observations' versus models

Avg annual precipitation intensity (mm/d) over land (masked)







Remember that observations are models

- "All models are wrong..."
- "..but SOME are useful"

- "All models using erroneous data are wrong..."
- "...and NONE are useful"



G.E.P. Box



L.V. Alexander